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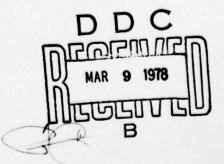


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University of Kentucky Department of Statistics





CHARACTERIZATIONS OF GEOMETRIC DISTRIBUTION AND DISCRETE IFR (DFR) DISTRIBUTIONS USING ORDER STATISTICS

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Abstract

Let X be a discrete random variable the set of possible values (finite or infinite) of which can be arranged as an increasing sequence of real numbers $a_1 < a_2 < a_3 < \dots$ In particular, a_i could be equal to i for all i. Let $X_{1n} \le X_{2n} \le \dots \le X_{nn}$ denote the order statistics in a random sample of size n drawn from the distribution of X, where n is a fixed integer ≥ 2 . Then, we show that for some arbitrarily fixed $k(2 \le k \le n)$, independence of the event $\{X_{kn} = X_{ln}\}$ and X_{ln} is equivalent to X being either degenerate or geometric. We also show that the montonicity in i of $P\{X_{kn}=X_{ln}|X_{ln}=a_i\}$ is equivalent to X having the IFR (DFR) property. Let $a_i = i$ and $G(i) = P(X \ge i)$, i = 1, 2, ... We prove that the independence of $\{X_{2n}-X_{1n} \in B\}$ and X_{1n} for all i is equivalent to X being geometric, where $B=\{m\}$ ($B=\{m,m+1,...\}$), provided $G(i)=q^{i-1}$, $1 \le i \le m+2$ (1 < i < m+1), where 0 < q < 1.

Introduction.

Several contributions have been made to characterizing the geometric distribution using order statistics. Ferguson (1965) has shown that the independence of the smallest observation and the sample range in a random sample of size 2 drawn from a non-degenerate discrete population implies and is implied by the discrete distribution being geometric. If the underlying distribution is that of an unbounded lattice variate, Srivastava (1974) has shown that X_{1n} and the event $\{X_{1n} = \dots = X_{nn}\}$ are independent if and only if the distribution is geometric, where X denotes the ith smallest order statistic in a random sample of size n (i=1,...,n). Galambos (1975) has extended Srivastava's result to the situation where the set of possible values of the discrete random DISTRIBUTION/AVAILABILITY CODES

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variable (finite or infinite) can be arranged in an increasing sequence (i.e. the set of possible values need not be of the form $\{\alpha+\beta i,\ i=1,2,\ldots,\beta\neq 0\}$. The main theme of our paper is to generalize the existing results in two directions: (i) For some arbitrarily fixed $k(2 \le k \le n)$ the independence of X_{ln} and $\{X_{kn}=X_{ln}\}$ should suffice to characterize the geometric distribution. (ii) For $a_i=1$, the independence of X_{ln} and $\{X_{2n}-X_{ln}=m\}$, or X_{ln} and $\{X_{2n}-X_{ln}\geq m\}$ for some fixed $m\geq 1$ should suffice to characterize the geometric distribution. In addition, monotonicity of $P(X_{kn}=X_{ln}|X_{ln}=a_i)$ in i for some arbitrarily fixed k can be employed to characterize the discrete IFR (DFR) distributions.

2. Notation and Definitions.

The random variables $X_{1n} \leq X_{2n} \leq \ldots \leq X_{nn}$ denote the order statistics corresponding to the n i.i.d. random variable X_1, \ldots, X_n . We denote by N the set of the natural numbers and by I a segment of N, where by a segment we mean that either I = N, or I = {i \(\in N \): i \(< r \)} for some reN.

The sequence of real numbers $\{a_i: i \in I\}$ is said to be strictly increasing if $a_i < a_j$ when i < j, $i,j \in I$.

Throughout this paper "increasing" is used in place of "nondecreasing" and "decreasing" is used in place of "nonincreasing".

<u>Definition 2.1.</u> Let X be a discrete random variable the set of possible values of which can be represented by a strictly increasing sequence of real numbers $\{a_i: i\in I\}$. Let $G(i)=P(X a_i)$, $i\in I$. Then X is said to have increasing (decreasing) failure rate distribution (denoted by IFR (DFR) distribution), if G(i+1)/G(i) decreases (increases) in $i\in I$.

<u>Definition 2.2.</u> Let X be a discrete random variable the set of possible values of which can be represented by a strictly increasing infinite sequence of real number $a_1 < a_2 < \ldots$ Let $G(i)=P(X \ge a_1)$, $i=1,2,\ldots$ The random variable X is said to be geometric if $G(i)=q^{i-1}$, $i=1,2,\ldots$, where 0 < q < 1.

3. Main Results.

Let X_1 , X_2 , ..., X_n , $n\geq 2$ be independent and identically distributed (i.i.d.) discrete random variables. Assume that the set of possible values of X_1 can be represented by a strictly increasing sequence of real numbers $\{a_i: i\in I\}$. In particular, a_i could be equal to i for all i.

The following Lemma gives a characterization of degenerate random variables and is useful in proving Theorem 3.1.

<u>Lemma 3.1</u>. Let X_1 be a discrete random variable. Then X_1 is degenerate if and only if $P(X_{1n} = X_{kn}) = 1$, where k is an arbitrarily fixed postive integer $(2 \le k \le n)$.

<u>Proof.</u> If X_1 is degenerate then trivially $P(X_{1n}=X_{kn})=1$. Now assume that X_1 is non-degenerate. Then there exists two real numbers b_1 , b_2 such that $P(X_1=b_1)>0$ and $P(X_1=b_2)>0$, where, without loss of generality, we assume that $b_1 < b_2$. Now, $P\{X_{1n}\neq X_{kn}\} \ge P\{X_{1n}=b_1, X_{2n}=X_{3n}=\dots=X_{nn}=b_2\}>0$, therefore $P\{X_{1n}=X_{kn}\}<1$ which completes the proof.

Remark 3.1. It should be noted that the conclusion of Lemma 3.1 remains valid even if X_1 is an arbitrary random variable.

We are ready to state and prove the main results.

Theorem 3.1. Let X_1 be a discrete random variable the set of possible values of which can be represented by a strictly increasing sequence of real numbers $\{a_i\colon i\in I\}$. Let k be an arbitrarily fixed positive integer $(2\leq k \leq n)$. Then X_{1n} is independent of the event $\{X_{1n}=X_{kn}\}$ if and only if X_1 is degenerate or $P(X_1\geq a_1)=q^{1-1}$, $i=1,2,\ldots$, where 0< q<1.

<u>Proof.</u> First observe that if X_1 is degenerate or if $P(X \ge a_i) = q^{i-1}$, i = 1, 2, ..., then in either case X_{1n} is independent of the event $\{X_{1n} = X_{kn}\}$. Next, in order to prove the converse, let $G(i) = P(X \ge a_i)$. By hypothesis we have

$$\begin{split} & P(X_{kn} = X_{ln}, \ X_{ln} = a_i) = P(X_{kn} = X_{ln}) \ P(X_{ln} = a_i). \ \ \text{Writing } P(X_{ln} = X_{kn} = a_i) \\ & = \sum_{j=k}^{n} \binom{n}{j} \left[G(i) - G(i+1) \right]^{j} \left[G(i+1) \right]^{n-j}, \ \text{and setting } j' = n-j \ \text{we are led to the} \end{split}$$

following equation:

$$\sum_{j'=0}^{n-k} {n \choose j'} [G(i+1)]^{j'} [G(i)-G(i+1)]^{n-j'} = P(X_{1n} = X_{kn})[G^{n}(i)-G^{n}(i+1)],$$
for all isI. (3.1)

Now either $I = \{i \in \mathbb{N}: i \le r\}$ for some $r \in \mathbb{N}$ or $I = \mathbb{N}$. In case $I = \{i \in \mathbb{N}: i \le r\}$ for some $r \in \mathbb{N}$, then setting i = r in (3.1) we obtain

$$G^{n}(r) = P(X_{|n} = X_{kn})G^{n}(r)$$
 where $G(r) > 0$.

Hence we must have $P(X_{1n}=X_{kn})=1$, which by Lemma 3.1 implies that X_1 is degenerate. Next, assume that I=N. Dividing both sides in (3.1) by $G^{n}(i)$ and letting q(i) = G(i+1)/G(i) we have

$$\begin{cases} \sum_{j=0}^{n-k} {n \choose j} [q(i)]^{j} [1-q(i)]^{n-j} \} (1-[q(i)]^{n})^{-1} = P(X_{1n} = X_{kn}), \\ \text{for } i=1,2,.... \quad (3.2) \end{cases}$$

Notice that 0 < q(i) < 1. Let Y_i be a binomial random variable with parameters (n, q(i)), i=1,2,..., then the numerator of L.H.S. of (3.2) is $P(Y_i \le n-k)$. Since $P(Y_i \le n-k) = 1-P(Y_i \ge n-k+1) = k \binom{n}{k} \int_{q(i)}^{1} u^{n-k} (1-u)^{k-1} du$, the L.H.S. of (3.2) can be written as $\{k \binom{n}{k} \int_{0}^{1-q(i)} t^{k-1} (1-t)^{n-k} dt\}/(1-q^n(i))$. Now since the R.H.S. of (3.2) is free of i the L.H.S. is constant in i=1,2,3,... Now let $f(x) = \{k \binom{n}{k} \int_{0}^{1-x} t^{k-1} (1-t)^{n-k} dt\}/(1-x^n)$, 0 < x < 1. (3.3)

Differentiating with respect to x we have

$$f'(x) = \{k\binom{n}{k}x^{n-k} [nx^{k-1} \int_0^{1-x} t^{k-1} (1-t)^{n-k} dt - (1-x)^{k-1} (1-x^n)]\} (1-x^n)^{-2}.$$

To show that f'(x) < 0, 0 < x < 1, we first observe that

 $\begin{array}{l} nx^{k-1} \int_0^{1-x} t^{k-1} \left(1-t\right)^{n-k} dt \ - \left(1-x\right)^{k-1} \left(1-x^n\right) \ \leq \ \left(1-x\right)^{k-1} \left[nx^{k-1} \int_0^{1-x} \left(1-t\right)^{n-k} dt - \left(1-x^n\right)\right] \\ \\ = \ \frac{\left(1-x\right)^{k-1}}{n-k+1} \left[nx^{k-1} - \left(n-k+1\right) - \left(k-1\right)x^n\right]. \end{array}$

Now, let $g(x) = nx^{k-1} - (n-k+1) - (k-1)x^n$. Since g(0) < 0, g(1) = 0 and $g'(x) = n(k-1)x^{k-2}(1-x^{n-k+1}) > 0$ for 0 < x < 1 it follows that g(x) < 0 for 0 < x < 1.

Consequently f'(x) < 0, 0 < x < 1 which implies that f(x) is strictly decreasing. This together with (3.2) implies that q(i) is constant for $i=1,2,\ldots$. Let q(i)=q where 0 < q < 1. It follows that $G(i)=q^{i-1}$, $i=1,2,\ldots$, which completes the proof of the theorem.

The following is an easy corollary 2 Theorem 3.1:

Corollary 3.1.1. Let X_1 be as in Theorem 3.1. The X_{ln} is independent of the event $\{X_{kn} > X_{ln}\}$ if and only if X_1 is degenerate or $P(X_1 \ge a_1) = q^{1-1}$, i=1,2,..., 0 < q < 1.

<u>Proof.</u> The proof follows immediately by observing that the event $\{X_{kn} > X_{ln}\}$ is the complement of the event $\{X_{ln} = X_{kn}\}$.

Remark 3.1.1. Theorem 3.1 states that X_{1n} and $\{X_{1n} = \dots = X_{kn}\}$ are independent if and only if X_{1} has geometric distribution or X is degenerate. In particular, when k=n. Theorem 3.1 conincides with Galambos' (1975) result.

Our next theorem gives a characterization of the discrete IFR (DFR) distributions in terms of the montonicity in i of $P\{X_{ln}=X_{kn} | X_{ln}=a_i\}$. Such a characterization will be useful in constructing statistical tests for such classes of life distributions.

Theorem 3.2. Let X_1 be as in Theorem 3.1. Then X_1 has IFR (DFR) distribution if and only if $P\{X_{1n}=X_{kn} \mid X_{1n}=a_i\}$ increases (decreases) in i, where again $2 \le k \le n$ is an arbitrarily fixed integer.

Proof. As in the proof of Theorem 3.1 we have

 $P\{X_{1n}=X_{kn} \mid X_{1n}=a_i\} = \{k \ \binom{n}{k} \ \int_0^{1-q(i)} t^{k-1} \ (1-t)^{n-k} \ dt \} (1-q^n(i))^{-1}, \text{ iel }$ where q(i)=G(i+1)/G(i) iel. {Notice that G(i)>0 for iel}. Again let $f(x)=\{k \ \binom{n}{k} \ \int_0^{1-x} t^{k-1} \ (1-t)^{n-k} \ dt \} (1-x^n)^{-1}, \quad 0 \leq x \leq 1. \text{ We have shown in }$ the proof of Theorem 3.1 that f(x) is strictly decreasing in x. Consequently $P\{X_{1n}=X_{kn} \mid X_{1n}=a_i\} \text{ increases (decreases) in i if and only if } G(i+1)/G(i)$ decreases (increases) in i, which completes the proof.

Remark 3.2.1. One may give the following intuitive explanation of Theorem 3.2. If X_1 has an increasing failure rate then as the given value of X_{1n} gets larger, the values of $X_1, \ldots X_n$ are more likely to be "close" to one another. Consequently the probability of ties among X_{1n}, \ldots, X_{nn} gets higher. Similar intuitive explanations of Theorem 3.1 can be given that is based on the "lack of memory" property of the geometric distribution.

Let X_1 be as in Theorem 3.1, and assume that $a_i=i$, $i\in I$. Then for k=2, Theorem 3.1 can be stated as follows: X_{1n} is independent of $\{X_{2n}-X_{1n}=0\}$ if and only if X_1 is degenerate or $P(X_1\ge i)=q^{i-1}$, $i=1,2,\ldots,\ 0< q<1$. One might ask whether the event $\{X_{2n}-X_{1n}=0\}$ can be replaced by the event $\{X_{2n}-X_{1n}=m\}$ or $\{X_{2n}-X_{1n}\ge m\}$ where m>0? The following theorem gives an affirmative answer provided we assume some boundary conditions (which automatically rule out the possibility of X_1 being degenerate).

Theorem 3.3. Let X_1 be a discrete random variable the set of possible values of which is I. Let $G(i) = P(X \ge i)$, isI, and $m \ge 1$ be arbitrarily fixed positive integer. Then

- (i) $G(i) = q^{i-1} \ 1 \le i \le m+2 \ 0 < q < 1 \ and \ X_{1n}$ is independent of the event $\{X_{2n} X_{1n} = m\}$ if and only if $G(i) = q^{i-1}$, i=1,2,3,...
- (ii) $G(i) = q^{i-1}$, $i \le 1 \le m+1$, 0 < q < 1, and X_{1n} is independent of the event $\{X_{2n} X_{1n} \ge m\}$ if and only if $G(i) = q^{i-1}$, i=1,2,...

<u>Proof.</u> We provide the proof for (ii) only, since (i) can be proved in a similar fashion. By the independence assumption we have

$$P(X_{2n}-X_{1n})=m \mid X_{1n}=i)$$
 is free of i, where iEI. (3.4)

Now

$$P(X_{2n} - X_{1n} - x_{1n}) = [P(X_{2n} - x_{1n} - x_{1n}) - P(X_{2n} - x_{1n} - x_{1n})] / [P(X_{1n} - x_{1n})] / [P(X_{1n} - x_{1n} - x_{1n})] / [P(X_{1n} - x_{1n})] /$$

Setting i=1 and using (3.4) we have

$$(nG^{n-1}(1+m)[G(1)-G(2)])/(G^{n}(1)-G^{n}(2)) = (nG^{n-1}(i+m)[G(i)-G(i+1)])/(G^{n}(i)-G^{n}(i+1))$$
(3.5)

By the boundary conditions the L.H.S. of (3.5) is equal to $(n \ q^{(n-1)m}[1-q])/(1-q^n).$ Substituting in (3.5) and using induction we obtain $G(i) = q^{i-1}$, $i=1,2,\ldots$, i.e. X_1 is geometric and the proof is now complete.

Remark 3.3.1. Notice that results (i) and (ii) in Theorem 3.3 have different sets of boundary conditions. Also notice that for m=1, (ii) is subsumed by Corollary 3.1.1 with k=2 and $a_i=i$.

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